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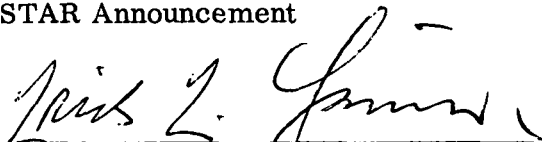
ITERATIVE COMPUTATION OF GENERALIZED  
INVERSES, WITH AN APPLICATION TO CMG  
STEERING LAWS

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Program Development

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## LIST OF SYMBOLS

<u>Symbol</u>	<u>Definition</u>
$A = [a_{ij}]$	$M \times N$ matrix $A$
$A^*$	$[\bar{a}_{ji}]$ , an $N \times M$ matrix
$A^+$	(Moore-Penrose) generalized inverse of the matrix $A$
$I$	the identity matrix
$\lambda_1(B)$	largest eigenvalue of the positive semidefinite matrix $B$
$\ A\ $	$\lambda_1(AA^*)$ , the spherical matrix norm, or the euclidian vector norm
$P_{R(A)}$	$AA^+$ , the projection onto the range space of $A$
$P_{R(A^*)}$	$A^+A$ , the projection onto the range space of $A^*$

# ITERATIVE COMPUTATION OF GENERALIZED INVERSES WITH AN APPLICATION TO CMG STEERING LAWS

## INTRODUCTION

The equation

$$Ax = b$$

will possess a solution if, and only if,  $b$  lies in the coordinate space spanned by the columns of  $A$ , i.e., the range space  $R(A)$ , and in this case every solution can be put in the form

$$x_g = A^+ b + (I - A^+ A) z$$

where  $z$  is some  $N \times 1$  column. That solution with  $z = 0$ , denoted by  $x_b$ , has the property that for all  $z$ ,

$$\|x_b\| \leq \|x_g\|$$

i.e.,  $x_b$  is the minimum-norm solution. If no solution exists, then there is some vector in  $R(A)$  which is closest to  $b$  in the least-square sense; this vector is just  $x_b = A^+ b$  so that for all  $x$

$$\|b - Ax_b\| \leq \|b - Ax\|$$

These properties establish the importance of the generalized inverse for applications. Unfortunately, there is no truly convenient generally applicable formula for its computation. Various elimination algorithms based upon the defining equation, i.e.,

$$\begin{aligned} AXA &= A \\ XAX &= X \\ (AX)^* &= AX \\ (XA)^* &= XA \end{aligned}$$

have been devised, but these are sensitive to the preliminary computation of the rank of A. This report describes an iterative scheme for computation of the generalized inverse and incorporates the scheme into a FORTRAN subroutine which may also be used for computing the inverses of nonsingular square matrices.

## FORTRAN SUBROUTINE DESCRIPTION

The Fortran subroutine listed below is called by the following statement:

CALL GNVERS (Y0, T, R, M, N, A, BD, AP)

where A is an  $M \times N$  matrix; BD is any number not less than  $\lambda_1(AA^*)$ ; and AP, the output, is the  $N \times M$  generalized inverse of A. Y0, T, R are matrices used by GNVERS and need not be defined by the user. A convenient value for BD is

$$\max_{1 \leq i \leq M} \sum_{j=1}^M |b_{ij}|$$

where

$$B = \begin{bmatrix} b_{ij} \end{bmatrix}_1^M = AA^* \quad .$$

The calling program must contain the following statement:

DIMENSION Y0( ), T( ), R( ) .

The arguments in the dimension statement must not be less than the following numbers:  $N \times M$ ,  $M \times M$ ,  $M \times M$ , respectively. It is assumed that  $M \leq N$ . If  $M > N$ , the generalized inverse of  $A^*$  should be computed: This stratagem is adopted to keep the dimensions of certain intermediate matrices small (thereby reducing the number of arithmetical operations and computation time) and is acceptable since  $(A^*)^+ = (A^+)^*$  .

The algorithm implemented via this subroutine forms successive iterates  $Y_1, Y_2, \dots$  which will converge to  $A^+$ : The maximum number of

iterations is denoted by ITERs, and is set to 20 in the listing below. The iterations cease when

$$E \equiv \max_{\substack{1 \leq i \leq N \\ 1 \leq j \leq M}} \left\{ \left| y_{ij}^{(n)} - y_{ij}^{(n-1)} \right| \right\} < \text{TOL} \quad (n = 2, 3, \dots)$$

where TOL is set to  $10^{-7}$  in the listing below. If convergence does not occur in ITERs iterations, an error statement is printed and the subroutine returns the last iterate to the calling program.

GNVERS makes use of a matrix-multiply subroutine whose calling statement is

CALL FMXMT (M, N, IP, A, B, AB)

which forms the  $M \times IP$  product AB of the  $M \times N$  matrix A and the  $N \times IP$  matrix B. A listing of this subroutine is also given.

The accuracy of the GNVERS subroutine is illustrated by the following example. The  $3 \times 4$  matrix A was defined as

$$A = \begin{bmatrix} 4.604359873-01 & 6.981586633-01 & 1.637202877-01 & 7.932543322-01 \\ 8.176181213-01 & -5.241646385-01 & 9.788190850-01 & 5.955607116-01 \\ 3.456868410-01 & -4.876741769-01 & -1.229181702-01 & -1.267093084-01 \end{bmatrix}.$$

A careful computation using an electronic desk calculator gave the following result:

$$A^+ = \begin{bmatrix} 7.030320327-01 & -6.471307646-02 & 1.493286880+00 \\ 5.203275028-01 & -2.903730234-01 & -5.949098340-01 \\ -6.275139883-01 & 8.483851012-01 & -1.552037349+00 \\ 5.241249280-01 & 1.180262294-01 & -2.284458009-02 \end{bmatrix},$$

while the subroutine GNVERS gave

$$A^+ = \begin{bmatrix} 7.03032032-01 & -6.47130762-02 & 1.49328688+00 \\ 5.20327502-01 & -2.90373023-01 & -5.94909835-01 \\ -6.27513987-01 & 8.48385101-01 & -1.55203735+00 \\ 5.24124927-01 & 1.18026230-01 & -2.28445806-02 \end{bmatrix}.$$

## PROGRAM LISTING

The main subroutine listing is given below.

```
SUBROUTINE GNVERS(Y0,T,R,M,N,A,BD,AP)
DIMENSION A(1),Y0(1),Y1(1),T(1),R(1)

ITERS=20
TOL=1.E-07
MN=M*N
MSQR=M*M
ALF=1.6/BD
DO1 I=1,N
DO1 J=1,M
INDEX1=(J-1)*N+I
INDEX2=(I-1)*M+J
1  Y0(INDEX1)=ALF*A(INDEX2)
DO2 L=1,ITERS
CALL FMXMT(M,N,M,A,Y0,R)
DO3 I=1,MSQR
3  R(I)=-R(I)
DO4 I=1,M
INDEX1=(J-1)*M+I
4  R(INDEX1)=1.0+R(INDEX1)
DO5 I=1,MSQR
5  Y1(I)=R(I)
DO6 I=1,M
INDEX1=(I-1)*M+I
6  Y1(INDEX1)=1.0+Y1(INDEX1)
CALL FMXMT(M,M,M,Y1,R,T)
DO7 I=1,M
INDEX1=(I-1)*M+I
7  T(INDEX1)=1.0+T(INDEX1)
CALL FMXMT(N,M,M,Y0,T,Y1)
E=0.0
DO8 I=1,MN
Y=ABSF(Y1(I)-Y0(I))
IF(Y.GT.E)9,8
9  E=Y
8  CONTINUE
1F(E.LT.TOL)99,10
```



```

10 DO2 I=1, MN
  2 Y0(I)=Y1(I)
    PRINT11, ITERS
11 FORMAT(1X, 28HGNVERS DOES NOT CONVERGE IN, I3, 11H ITERATIONS)
99 CONTINUE
RETURN
END

```

The FMXMT listing is

```

SUBROUTINE FMXMT(M, N, IP, A, B, AB)
DIMENSION A(1), B(1), AB(1)
DO1 I=1, M
DO1 J=1, IP
L=(J-1)*M+J
AB(L)=0.0
DO1 K=1, N
K1=(K-1)*M+I
K2=(J-1)*N+K
1 AB(L)=AB(L)+A(K1)*B(K2)
RETURN
END

```

## DERIVATION OF THE ALGORITHM

The Schulz method for inversion of a nonsingular matrix is discussed in Reference 1 and 2. The generalized form of this algorithm, called the hyperpower method of degree  $m$ , is defined by

$$\left. \begin{aligned} R_n &= I - AX_n \\ X_{n+1} &= X_n \left( I + R_n + R_n^2 + \dots + R_n^{m-1} \right) \end{aligned} \right\} n = 0, 1, 2, \dots$$

In Reference 1 it is shown that the hyperpower method of degree 3 is optimum in a certain sense, and convergence bounds are given. References 3, 4, and 5 discuss the hyperpower method of degree 2 (the usual form of Schulz's algorithm) applied to the problem of computing the generalized inverse of a matrix  $A$ . Convergence bounds and optimal initialization values are investigated in References 1 and 5. Following the techniques in Reference 4 Lemma 1 and Corollary 1 establish the hyperpower method of degree  $m$  for generalized inversion.

Lemma 1: The sequence of matrices defined by

$$\left. \begin{aligned} R_n &= P_{R(A)} - AX_n \\ X_{n+1} &= X_n \left( P_{R(A)} + R_n + R_n^2 + \dots + R_n^{m-1} \right) \end{aligned} \right\} n = 0, 1, 2, \dots \quad (1)$$

$$(2)$$

where

$$X_0 = A^* B_0 \quad (B_0 \text{ is some nonsingular } M \times M \text{ matrix}) \quad , \quad (3)$$

$$X_0 = C_0 A^* \quad (C_0 \text{ is some nonsingular } N \times N \text{ matrix}) \quad , \quad (4)$$

$$\| P_{R(A)} - AX_0 \| < 1 \quad , \quad (5)$$

and

$$\| P_{R(A^*)} - X_0 A \| < 1 \quad . \quad (6)$$

converges to  $A^+$  as  $n \rightarrow \infty$ .

**Proof:** Since  $A^+$  is the unique solution of the equations

$$AX = P_{R(A)} \quad (7)$$

and

$$XA = P_{R(A^*)} \quad , \quad (8)$$

it suffices to prove that

$$\lim_{n \rightarrow \infty} \| P_{R(A)} - AX_n \| = 0 \quad (9)$$

$$\lim_{n \rightarrow \infty} \| P_{R(A^*)} - X_n A \| = 0 \quad (10)$$

To do this, consider the sequences of matrices

$$B_{n+1} = B_n \left( P_{R(A)} + P_n + P_n^2 + \dots + P_n^{m-1} \right)$$

$$C_{n+1} = C_n \left( P_{R(A^*)} + Q_n + Q_n^2 + \dots + Q_n^{m-1} \right)$$

where

$$P_n = P_{R(A)} - AA^*B_n, \quad ,$$

$$Q_n = P_{R(A^*)} - A^*AC_n, \quad ,$$

and  $(n = 0, 1, 2, \dots)$ . Then it is obvious that

$$X_n = A^*B_n \quad (11)$$

Moreover,

$$X_n = C_n A^* \quad (12)$$

because if we let

$$Y_{n+1} = C_{n+1} A^*$$

and note that

$$\begin{aligned} Q_n^i A^* &= Q_n^{i-1} \left( P_{R(A^*)} A^* - A^* A C_n A^* \right) = Q_n^{i-1} A^* \left( P_{R(A)} - A Y_n \right) \\ &= \dots = A^* \left( P_{R(A)} - A Y_n \right)^i, \end{aligned}$$

since

$$P_{R(A^*)} A^* = A^* P_{R(A)}, \quad ,$$

then

$$Y_{n+1} = Y_n \left[ P_{R(A)} + (P_{R(A)} - AY_n) + (P_{R(A)} - AY_n)^2 + \dots + (P_{R(A)} - AY_n)^{m-1} \right]$$

so that  $Y_n$  satisfies equation (2). Since  $X_0 = Y_0$ ,  $X_n = Y_n$  for all  $n$ , which establishes equation (12).

Since

$$AX_n = P_{R(A)} AX_n = AX_n P_{R(A)},$$

we have by an easy induction that

$$R_n^k = P_{R(A)} - AX_n \left( P_{R(A)} + R_n + R_n^2 + \dots + R_n^{k-1} \right) \\ (k = 0, 1, 2, \dots)$$

Setting  $k = m$  gives  $R_{n+1} = R_n^m$ , so that

$$\|P_{R(A)} - AX_{n+1}\| \leq \|P_{R(A)} - AX_n\|^m,$$

which with equation (5) proves equation (9). The proof of equation (10) is analogous.

Corollary 1: Let

$$\lambda_1(AA^*) \geq \lambda_2(AA^*) \geq \dots \geq \lambda_r(AA^*)$$

denote the nonzero eigenvalues of  $AA^*$ . If

$$0 < \alpha < \frac{2}{\lambda_1(AA^*)}, \quad (13)$$

then the sequence defined by

$$Y_0 = \alpha A^* \quad (14)$$

$$R_n = I - AY_n \quad (15)$$

$$Y_{n+1} = Y_n \left( I + R_n + R_n^2 + \dots + R_n^{m-1} \right) \quad (16)$$

converges to  $A$  as  $n \rightarrow \infty$ .

Proof: Since

$$P_{R(A)} AA^* = AA^+ AA^* = AA^* = AA^* AA^+ = AA^* P_{R(A)} ,$$

$P_{R(A)}$  and  $AA^*$  are commuting hermitian matrices with the same range space, so that the eigenvalues of  $P_{R(A)} - AY_0$  are

$$\left. \begin{array}{ll} 1 - \alpha \lambda_i(AA^*) & (i = 1, 2, \dots, r) \\ 0 & (i = r+1, \dots, m) \end{array} \right\} ; \quad (17)$$

hence, by equation (13)

$$\left| \lambda_1(P_{R(A)} - AY_0) \right| < 1 \quad (18)$$

Similarly

$$\left| \lambda_1(P_{R(A^*)} - Y_0 A) \right| < 1 .$$

Now the process (2), begun with equation (14), retains the form of equation (12), so

$$\begin{aligned} X_n(P_{R(A)} + R_n + \dots + R_n^{m-1}) &= X_n P_{R(A)} + X_n(R_n + \dots + R_n^{m-1}) \\ &= C_n A^* AA^+ + X_n(R_n + \dots + R_n^{m-1}) \\ &= X_n(I + R_n + \dots + R_n^{m-1}) \end{aligned}$$

Thus, the convergence of equation (16) follows from that of equation (12).

Corollary 2: The process (18) is convergent if  $\alpha$  satisfies the inequality

$$0 < \alpha < 1 \leq \max_{1 \leq i \leq M} \sum_{j=1}^M |b_{ij}|$$

where  $B = \begin{bmatrix} b_{ij} \end{bmatrix}_1^M = AA^*$ .

Proof: By Gershgorin's theorem [ 2 ],

$$\lambda_1(AA^*) \leq \max_{1 \leq i \leq M} \sum_{j=1}^M |b_{ij}|$$

Note that it is immaterial whether we take B as  $AA^*$  or  $A^*A$  since the nonzero eigenvalues of these two matrices are the same.

## APPLICATION TO REDUNDANT CMG ASSEMBLIES

The instantaneous torque output of a single-gimbal CMG is given by

$$\dot{h} = \dot{\alpha}_1 e_1 \times h e_2 = h \dot{\alpha} e$$

where  $\dot{\alpha}$  is the instantaneous gimbal rate,  $h$  is the CMG momentum magnitude (determined by the wheel speed), and  $e_1$  and  $e_2$  are unit vectors along the gimbal axis and in the direction of the momentum vector, respectively. Then  $e$  is a unit vector along the instantaneous torque vector  $\dot{h}$ . The torque output of  $N$  such CMGs is then

$$\dot{h} = \dot{h}_1 + \dot{h}_2 + \dots + \dot{h}_N = h_1 \dot{\alpha}_1 e_1 + h_2 \dot{\alpha}_2 e_2 + \dots + h_N \dot{\alpha}_N e_N$$

where  $\dot{\alpha}_i$  denotes the gimbal rate of the  $i$ th CMG and  $e_i$  now denotes the unit vector in the direction of its instantaneous torque vector. This last equation can be put in the following form:

$$\dot{h} = C \dot{\alpha}$$

where  $C$  is the  $3 \times N$  matrix whose columns are  $h_1 e_1, h_2 e_2, \dots, h_N e_N$  and  $\dot{\alpha}$  is the column of gimbal rates. Then, we have that

$$B = A^* A = \begin{bmatrix} h_1^2 e_1^* e_1 & h_1 h_2 e_1^* e_2 & \dots & h_1 h_N e_1^* e_N \\ h_2 h_1 e_2^* e_1 & h_2^2 e_2^* e_2 & \dots & h_2 h_N e_2^* e_N \\ \dots & \dots & \dots & \dots \\ h_N h_1 e_N^* e_1 & h_N h_2 e_N^* e_2 & \dots & h_N^2 e_N^* e_N \end{bmatrix}$$

so

$$\lambda_1(A^* A) \leq \max_{1 \leq i \leq N} \sum_{j=1}^N h_i h_j |e_i^* e_j| \leq \max_{1 \leq i \leq N} \sum_{j=1}^N h_i h_j \leq N h^2$$

since  $|e_i^* e_j| \leq 1$  for each  $i$  and  $j$ ;  $h$  is the nominal momentum of the CMGs used.

If  $T_c$  is a commanded torque output, we wish to ensure that  $\dot{h} = T_c$  by choosing  $\dot{\alpha}$ ; the procedure by which this choice is made is called a steering law. One such law, which is optimum in a certain sense, is

$$\dot{\alpha} = C^+ T_c$$

While this steering law has exhibited satisfactory performance in a number of simulations, a number of complicating factors must be considered before the law can be considered for implementation. Discussion of these, which have mainly to do with situations in which the columns  $e_i$  become coplanar (a hangup condition), is beyond the scope of this report.

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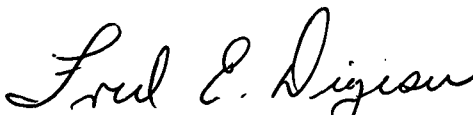
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By J. W. Steincamp

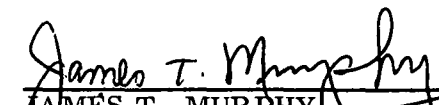
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